

PLS-1 Example

Copyright 2022-2023 Battelle Memorial Institute

This example shows how to compute Partial Least Squares (PLS) predictions for a single constituent. Computing PLS for a single constituent is known as PLS-1.

PLS can be computed with and without pre-processing. For PLS, pre-processing consists of mean-centering, which means to subtract out the mean from the data. This example shows how to compute PLS-1 with and without mean-centering.

PLS Algorithm

The PLS Algorithm is encapsulated in the following MATLAB function. If meanCentered is not entered, or if it is false, then pre-processing (mean centering) is not done. If meanCentered is true, then pre-processing (mean centering) is done.

```
function [C_pls, B_pls] = pnnl_pls(A_train, C_train, A_unknown, nLatentVariables, meanCentered)
    %pnnl_pls Partial least squares (PLS) regression
    %
    % [C_pls, B_pls] = pnnl_pls(A_train, C_train, A_unknown, nLatentVariables) returns
    % concentration matrix C_pls computed using the weights and loadings
    % from the SIMPLS algorithm on mean-centered A_train and C_train.
    % The relationship between multiplier matrix B_pls and C_pls is
    %  $C_{pls} = (A_{unknown} - \text{mean}(A_{train},1)) * B_{pls} + \text{mean}(C_{train},1)$ .
    %
    % pnnl_pls(A_train, C_train, A_unknown, nLatentVariables, meanCentered) applies mean
    % centering when meanCentered is true, and does not apply mean
    % centering when meanCentered is false. When meanCentered is not
    % supplied, the default is false (no mean centering).
    %
    % Example:
    %
    % load pnnl_napalm_data
    % nLatentVariables = 3;
    % meanCentered = true;
    % [C_pls, B_pls] = pnnl_pls(A_train, C_train, ...
    %                             A_unknown, nLatentVariables, meanCentered);
    %
    % See also pnnl_cls, pnnl_pcr.

    % Copyright 2022-2023 Battelle Memorial Institute
    if nargin < 5
        meanCentered = false;
    end

    X = A_train;
    Y = C_train;
    if meanCentered
        X0 = X - mean(X,1);
        Y0 = Y - mean(Y,1);
    else
        X0 = X;
        Y0 = Y;
    end
end
```

```

[X_loadings,Y_loadings,X_scores,Y_scores,Weights] = pnnl_simpls(X0,Y0,nLatentVariables); %#ok<AS
B_pls = Weights * Y_loadings';

if meanCentered
    C_pls = (A_unknown - mean(A_train,1)) * B_pls + mean(C_train,1);
else
    C_pls = A_unknown * B_pls;
end
end

```

Concentration Data

The concentrations of the training data are in matrix C_{train} and the concentrations of the validation data are in matrix $C_{\text{validation}}$. Column 1 corresponds to the concentrations in constituent 1 (benzene). Column 2 corresponds to the concentrations in constituent 2 (polystyrene). Column 3 corresponds to the concentrations in constituent 3 (gasoline).

$$C_{\text{train}} = \begin{array}{c} \begin{array}{ccc} \textit{benzene} & \textit{polystyrene} & \textit{gasoline} \end{array} \\ \begin{bmatrix} 0 & 0 & 100.0000 \\ 5.1309 & 0 & 94.8691 \\ 10.0660 & 0 & 89.9300 \\ 20.1799 & 0 & 79.8201 \\ 40.0120 & 0 & 59.9878 \\ 59.9972 & 0 & 40.0028 \\ 79.8412 & 0 & 20.1588 \\ 89.8273 & 0 & 10.1727 \\ 100.0000 & 0 & 0 \\ 90.0264 & 9.9736 & 0 \\ 80.1375 & 19.8625 & 0 \\ 64.9950 & 35.0005 & 0 \\ 21.0228 & 45.9197 & 33.0575 \\ 49.9507 & 5.0599 & 44.9895 \\ 40.0182 & 20.0385 & 39.9433 \\ 40.0154 & 10.0036 & 49.9810 \\ 30.0059 & 10.0282 & 59.9659 \\ 40.0340 & 39.9670 & 19.9990 \\ 49.9393 & 3.3748 & 46.6859 \\ 46.6501 & 13.4658 & 39.8840 \end{bmatrix} \end{array} \begin{array}{c} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 17 \\ 18 \\ 19 \\ 20 \end{array}$$

$$C_{\text{validation}} = \begin{array}{c} \begin{array}{ccc} \textit{benzene} & \textit{polystyrene} & \textit{gasoline} \end{array} \\ \begin{bmatrix} 22.1665 & 39.0384 & 38.7951 \\ 21.6874 & 37.0596 & 41.2530 \\ 22.1665 & 39.6980 & 38.1355 \\ 26.9575 & 40.3576 & 32.6849 \\ 25.9993 & 33.7616 & 40.2391 \\ 23.1247 & 37.0596 & 39.8157 \\ 22.6456 & 39.0384 & 38.3160 \\ 22.6456 & 39.0384 & 38.3160 \\ 27.4366 & 42.3364 & 30.2270 \\ 27.4366 & 37.7192 & 34.8442 \\ 26.4784 & 40.3576 & 33.1640 \\ 26.9575 & 37.7192 & 35.3233 \end{bmatrix} \end{array} \begin{array}{c} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \\ 11 \\ 12 \end{array}$$

Clear variables and load the PNNL napalm data.

```

clearvars
load pnnl_napalm_data

```

PLS-1 with and without pre-processing

Choose the number of latent variables.

```
nLatentVariables = 3;
```

Set meanCentered to true to indicate that pre-processing (mean centering) is done, and false to indicate that pre-processing (mean centering) is not done.

```
for meanCentered = [true, false]
```

Set up the plot title and color.

```
LogicalStr = {'Not ', ''};  
title_string = sprintf('PLS-1, %sMean Centered, %d Latent Variables', LogicalStr{meanCentered}, nLatentVariables);  
nConstituents = size(C_validation, 2);  
colorOrder = pnnl_colorOrder(nConstituents);
```

For each of the three constituents, use the corresponding column of C_train to compute PLS. Computing PLS for a single constituent is known as PLS-1. Use the corresponding column of C_validation to compute RMSEP (root mean square error predicted). Use the corresponding column of C_train to compute RMSEC (root mean square error calibration) and RMSECV (root mean square error cross validation).

```
figure  
h = gobjects(nConstituents, 1);  
for k = 1:nConstituents  
    % Compute PLS  
    C_predicted = pnnl_pls(A_train, C_train(:, k), A_unknown, nLatentVariables, meanCentered);  
    C_calibration = pnnl_pls(A_train, C_train(:, k), A_train, nLatentVariables, meanCentered);  
    C_cross_validation = pnnl_cross_validation(@pnnl_pls, A_train, C_train(:, k), nLatentVariables, meanCentered);  
  
    % Compute RMSE  
    RMSEP = pnnl_rmse(C_validation(:, k), C_predicted);  
    RMSEC = pnnl_rmse(C_train(:, k), C_calibration);  
    RMSECV = pnnl_rmse(C_train(:, k), C_cross_validation);  
    % Display RMSE  
    pnnl_display_rmse(title_string, ConstituentNames(k), ...  
        RMSEC, RMSECV, RMSEP);  
  
    % Plot Concentrations  
    hold on  
    % Validation vs. Predicted  
    h(k) = plot(C_validation(:, k), C_predicted, '.', 'MarkerSize', 35, 'Color', colorOrder(k));  
    % Train vs. Calibration  
    plot(C_train(:, k), C_calibration, 'o', 'MarkerSize', 10, 'LineWidth', 1, 'Color', colorOrder(k));  
    % Train vs. Cross Validation  
    plot(C_train(:, k), C_cross_validation, '+', 'MarkerSize', 10, 'LineWidth', 1, 'Color', colorOrder(k));  
    % 1-1 line  
    line(C_train(:, k), C_train(:, k), 'Color', 'k')  
    title(title_string)  
    xlabel(['Known constituent (' , ConcentrationUnits, ')'])  
    ylabel(['Predicted constituent (' , ConcentrationUnits, ')'])  
    set(gca, 'FontSize', 14)  
    box on  
    axis square  
    hold off
```

```

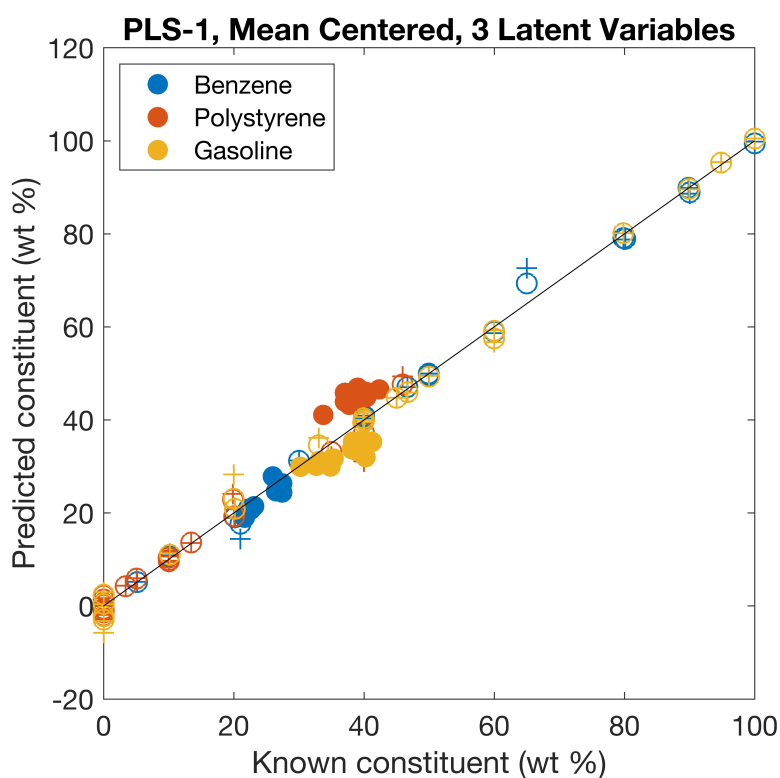
end
legend(h, 'Location', 'northwest')
disp('Legend: Dot is predicted. Circle is calibration. Cross is cross-validation.')
end

```

PLS-1, Mean Centered, 3 Latent Variables	Benzene
RMSEC	1.4067
RMSECV	2.3988
RMSEP	1.9121

PLS-1, Mean Centered, 3 Latent Variables	Polystyrene
RMSEC	1.4287
RMSECV	2.7267
RMSEP	6.4373

PLS-1, Mean Centered, 3 Latent Variables	Gasoline
RMSEC	1.4405
RMSECV	2.7035
RMSEP	4.475

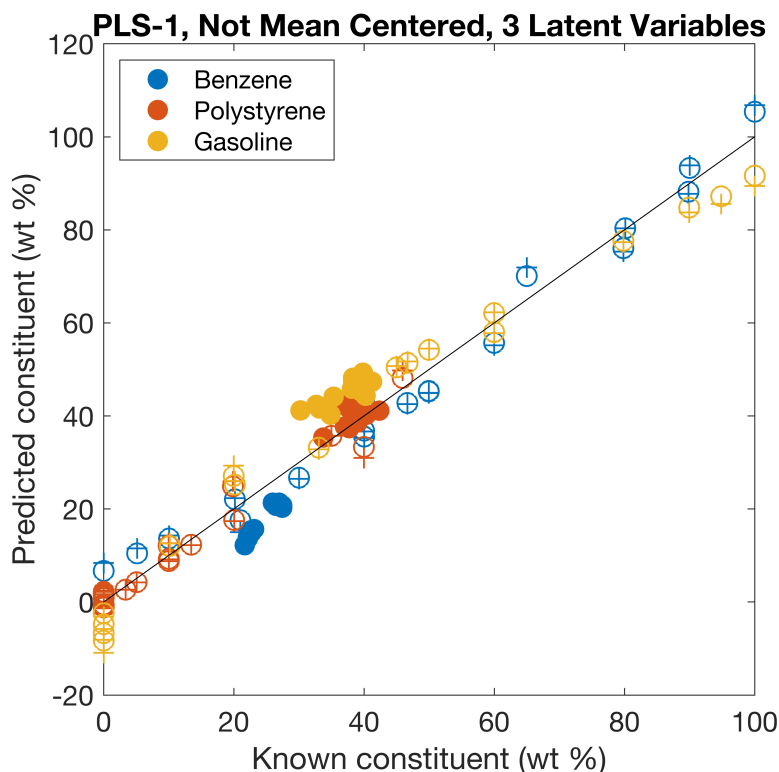


Legend: Dot is predicted. Circle is calibration. Cross is cross-validation.

PLS-1, Not Mean Centered, 3 Latent Variables	Benzene
RMSEC	3.9795
RMSECV	4.8301
RMSEP	7.189

PLS-1, Not Mean Centered, 3 Latent Variables	Polystyrene
RMSEC	2.2997
RMSECV	2.9529
RMSEP	2.3787

PLS-1, Not Mean Centered, 3 Latent Variables	Gasoline
RMSEC	5.2737
RMSECV	6.3277
RMSEP	8.3439



Legend: Dot is predicted. Circle is calibration. Cross is cross-validation.

Disclaimer

This material was prepared as an account of work sponsored by an agency of the United States Government. Neither the United States Government nor the United States Department of Energy, nor Battelle, nor any of their employees, nor any jurisdiction or organization that has cooperated in the development of these materials, makes any warranty, express or implied, or assumes any legal liability or responsibility for the accuracy, completeness, or usefulness or any information, apparatus, product, software, or process disclosed, or represents that its use would not infringe privately owned rights.

Reference herein to any specific commercial product, process, or service by trade name, trademark, manufacturer, or otherwise does not necessarily constitute or imply its endorsement, recommendation, or favoring by the United States Government or any agency thereof, or Battelle Memorial Institute. The views and opinions of authors expressed herein do not necessarily state or reflect those of the United States Government or any agency thereof.

PACIFIC NORTHWEST NATIONAL LABORATORY

operated by

BATTELLE

for the

UNITED STATES DEPARTMENT OF ENERGY

under Contract DE-AC05-76RL01830